## Assignment 5

## MATH 7502 - Semsester 2, 2018

## Mathematics for Data Science 2

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## Question 1

Suppose $G_{k+2}$ is the average of two previous numbers $G_{k+1}$ and $G_{k}$

$$
G_{k+2}=\frac{1}{2}\left(G_{k+1}+G_{k}\right), \quad G_{k+1}=G_{k+1}
$$

(a) Find the eigenvalues and eigenvectors of matrix $A$ such that

$$
\left[\begin{array}{c}
G_{k+2} \\
G_{k+1}
\end{array}\right]=A\left[\begin{array}{c}
G_{k+1} \\
G_{k}
\end{array}\right]
$$

(b) Find the limit as $n \rightarrow \infty$ of the matrices $A^{n}=X \Lambda X^{-1}$.
(c) If $G_{0}=0, G_{1}=1$, find the limit.

## Question 2

Let $A=\left[\begin{array}{ccc}6 & 3 & -2 \\ -4 & -1 & 2 \\ 13 & 9 & -3\end{array}\right]$.
(a) Evaluate $e^{t A}$.
(b) Find the general solutions of $\frac{d \vec{x}}{d t}=A \vec{x}$.
(c) Solve the initial value problem, if $\overrightarrow{x(0)}=\left[\begin{array}{c}-2 \\ 1 \\ 4\end{array}\right]$.
(d) Give an example of matrix $M$ and $N$ such that

$$
e^{M} e^{N} \neq e^{M+N}
$$

## Question 3

Find the singular value decomposition $U \Sigma V^{T}$ of $A$, where

$$
A=\left[\begin{array}{ccc}
3 & 2 & 2 \\
2 & 3 & -2
\end{array}\right]
$$

(a)What is the rank of $A$ ?
(b) Suggest two rank 1 approximations of $A$ based on the SVD. Which one is better?
(c) Write code that shows evidence of (b).

## Question 4

Suppose we take a sample of 7 from the students' tests in Math, Science and History, the result is the following:

| Math | Science | History |
| ---: | ---: | ---: |
| 7 | 5 | 4 |
| 2 | 1.5 | 2.0 |
| 3 | 9 | 8 |
| 3 | 7 | 7 |
| 5 | 3.5 | 5.0 |
| 6 | 4.5 | 5.0 |
| 7 | 3.5 | 4.5 |

(a) Compute the sample covariance matrix $S$.
(b) Find the eigenvalues of $S$.
(c) Write code for (a) and (b).
(d) Comment on the principal componenent analysis aspects of this and how they can be used.

## Question 5

Suppose $A x=\lambda x$. If $\lambda=0$ then $x$ is in the nullspace. If $\lambda \neq 0$ then $x$ is in the column space. Those spaces have dimensions $(n-r)+r=n$. Show that not every square matrix has $n$ linearly independent eigenvectors?

## Question 6

Consider a data-set with $2 M$ points, each a 2 -vector.
The first $M$ points are given by, $i=1, \ldots, M$ via the vectors $(i+1, i)$.
The later $M$ points are given by, $i=1, \ldots, M$ via the vectors $(i, i+1)$.
For example, for $M=10$ a plot of the points is given below.

In [62]:

```
using PyPlot
M = 10
points1 = [[i+1,i] for i in 1:M]
points2= [[i,i+1] for i in 1:M]
plot(first.(points1),last.(points1),".")
plot(first.(points2),last.(points2),".")
xlim(0,M+1);ylim(0,M+1);
```


(a) Give an explicit expression for the normalized covariance matrix of these points. As an aid, below is a numerical computation of the covariance matrix.

In [63]:

```
X = [reshape(vcat(points1...),2,M) reshape(vcat(points2...),2,M)]
muX = X*ones(2M, 2M)/2M
cvMat = (X-muX)*(X-muX)'/(2M)
```


## Out[63]:

$2 \times 2$ Array\{Float64,2\}:
8.58 .0
8.08 .5
(b) Carry out PCA for this data-set, numerically.
(c) Carry out PCA analytically and compare to your numerical results

## Question 7

Each pair of singular vectors $v$ and $u$ has $A v=\sigma u$ and $A^{T} u=\sigma v$.
(a) Show that the double vector $\left[\begin{array}{l}u \\ v\end{array}\right]$ is an eigenvector of the symmetric block matrix $M=\left[\begin{array}{cc}0 & A^{T} \\ A & 0\end{array}\right]$.
(b) Show that the SVD of $A$ is equivalent to the diagonalization of $M$.

## Question 8

Suppose the same $X$ diagonalizes both $A$ and $B$. They have same eigenvectors in $A=X \Lambda_{1} X^{-1}$ and $B=X \Lambda_{2} X^{-1}$. Prove that $A B=B A$.

## Question 9

Consider the code below.
(a) Describe in words what the code does.
(b) Replace line 5 of the code with a different implemination of pixel(). Have it create a square-looking donut. Carry out the computation with this image. Comment qualitatively on your results.

In [ ]:

```
using PyPlot,Distributions
```

$n, m=100,100$
sig $=0.1$
$\operatorname{pixel}(i, j)=\left((i-50)^{\wedge} 2+(j-50)^{\wedge} 2<25^{\wedge} 2 \& \&(i-50)^{\wedge} 2+(j-50)^{\wedge} 2>10^{\wedge} 2\right)+$
sig*rand(Normal())
$A=[p i x e l(i, j)$ for $i$ in $1: n, j$ in $1: m]$
$\mathrm{p}=\min (\mathrm{n}, \mathrm{m})$
$\mathrm{U}, \mathrm{S}, \mathrm{V}=\operatorname{svd}(\mathrm{A})$;
svdApprox(k) $=$ U[:, 1:k]*diagm(S[1:k])*V[:,1:k]'
err $=$ [norm(A-svdApprox(k)) for $k$ in 1:p]
plot(1:p,err)

In [ ]:

```
fig = figure(figsize=(10,10))
d = 15
for k in 1:d
    imm = fig[:add_subplot](d,1,k)
    imm[:imshow](svdApprox(k),cmap="Greys")
end
```


## Question 10

This is a carry over from Question 10 of Assignment 4. If you didn't do it there, do it now: Carry out 16.5 from [VMLS], page 352.

## Question 11

This is a carry over from Question 11 of Assignment 4. If you didn't do it there, do it now

## Question 12

This is a carry over from Question 12 of Assignment 4. If you didn't do it there, do it now

